THE RUNTIME DIFFERENCE BETWEEN THE A* AND DIJKSTRA'S **PATHFINDING ALGORITHMS IN SOLVING MAZE PROBLEMS**

Computer Science Extended Essay

Research Question

What is the difference between the runtime efficiency of Dijkstra's and the A* pathfinding algorithms in finding the shortest path in mazes with varying size?

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1. Introduction

Research Question: What is the difference between the runtime efficiency of Dijkstra's and the A* pathfinding algorithms in finding the shortest path in mazes with varying size?

Pathfinding algorithms (finding the shortest path between two set points on a grid), although might sound related only to technology, are an integral part of life. We, humans, have to determine our path in tasks like commuting to work, assessing the length and other factors of the road. Computers, however, need algorithms to determine the shortest path in such problems (Krafft 1,2). Pathfinding in computers is used in "navigation, video games, robotics, logistics" and others. (Algfoor, Sunar and Kolivand 1-3)

There are different pathfinding algorithms, from which Dijkstra's (Khan) and the A* algorithm (Mehta et al.) stand out as one of the most used algorithms.

This extended essay aims to investigate the runtime difference between Dijkstra's and the A* pathfinding algorithm in finding the shortest path from a starting and ending point in a maze problem with multiple paths between the starting and ending points in the maze.

This paper can aid especially in the video game and navigation fields. In real time strategy games such as Age of Empires, numerous units (armies, workers, etc.) constantly pathfind around in a large map consisting of a 256x256 grid (Cui and Shi 128,129). Even though the game is able to compute the paths of the units without visible lag, the players have consistently complained about units getting stuck or traverse a nonsensical path (H. Patel). Increasing the efficiency of the pathfinding algorithm used can aid in the better playability of the game by solving the existing problems.

Furthermore, autonomous drones are also starting to be a part of our lives, potentially shipping crucial cargo in the future. Employing the most efficient pathfinding algorithm can help in reducing costs and increasing mission success chances of such drones (Fu et al. 1,2). To investigate the difference between the algorithms, a maze generation algorithm (Recursive backtracking with dead end culling) along with the A* and Dijkstra's algorithms were programmed in C#. Their runtimes on different sizes of procedurally generated mazes were measured and analyzed.

2. Background Information

2.1 Procedural Maze Generation

Mazes are so old as to inspire Greek myths like the Minotaur and the labyrinth and are used currently as entertainment in means of video games (Pac-man, many roguelike games, etc.) or simply as puzzles to solve on the backs of newspapers (Hybesis). With the use of computers, completely random and very large mazes can be generated. There are many different procedural maze generation algorithms. (Pullen)

2.1.1 Maze Properties

Mazes have many different properties, indicating their nature. The properties relevant to this investigation are shown below.

Perfect mazes are defined by three properties: not having any passage loops, not having any isolated nodes and having only one path between any node pair in the maze. There are numerous ways to generate and solve such mazes as they are the most commonly used maze type. (Foltin 7)

Braided mazes, unlike perfect mazes, have no dead ends and may have multiple paths of varying length between two node in the maze (Foltin 7). Although there are different ways to generate such mazes (Ioannidis 31-35), the algorithms are much rarer since this maze type isn't as popular as perfect mazes.

Partial braided mazes are a combination of dead ends and loops. The ratio between the dead ends and loops can be calculated or manipulated. Similar to braided mazes, algorithms for the procedural generation for partial braided mazes are uncommon.

(Pullen)

The **elitism** of a maze is how much the solution of the maze covers its area. An elitist maze has a shorter and more direct solution, a non-elitist maze has a longer solution, covering more of its area. If there are multiple solutions, the elitism applies to the shortest path.

(Pullen)

2.1.2 The Recursive Backtracker Algorithm

A simple way to generate perfect mazes is the 'Recursive Backtracker' algorithm, which is based on the 'depth first search technique' (DFS). "The DFS algorithm wanders through the graph in a depth-oriented way". (Foltin 20-22) The algorithm travels whenever possible to a neighbor of the current node, and if it can't, it goes back to the previous vertex until it has iterated through all vertices. While generating a maze, a grid with node which all have 4 walls around them is firstly created. Then when the algorithm is traveling between vertices (or nodes), the wall between the two are destroyed, eventually generating a maze by boring walls though the grid. (Hybesis)

The algorithm has two possible implementations, either by recursion or iterative. (Ioannidis 23-25) The recursive version uses a lot of memory and is prone to overflow errors, while the iterative version uses a stack to store less data.

The steps for the iterative implementation and an illustration for the generation (figure 1) and a sample (figure 2) can be seen below.

- 1. Choose a starting point in the field.
- 2. Randomly choose a wall at that point and carve a passage through to the adjacent node, but only if the adjacent node has not been visited yet. This becomes the new current node.
- 3. If all adjacent node have been visited, back up to the last node that has uncarved walls (shown by the yellow points in figure 1) and repeat step 2.
- 4. The algorithm ends when the process has backed all the way up to the starting point. (Buck Maze Generation: Recursive Backtracking)

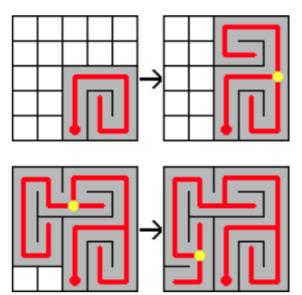


Figure 1: Image depicting recursive backtracker steps (Foltin 22)

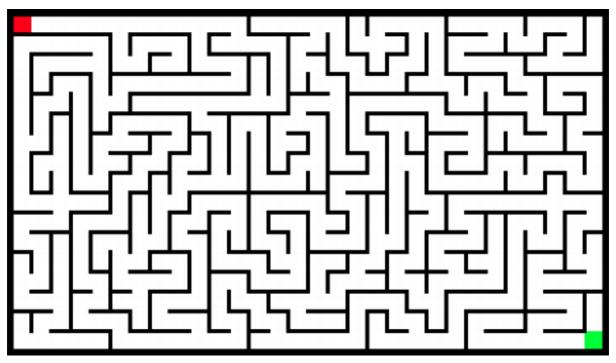


Figure 2: Image depicting a maze created using the recursive backtracking algorithm (Ioannidis 28)

2.1.3 Dead End Culling

While a vast number of algorithms exist for perfect maze generation, that is not the case for braided maze generation. Even though algorithms such as "Random Restarts" (Ioannidis 35-39) exist, they are uncommon. An easy way to obtain braided mazes is applying "dead-end culling" to a perfect maze, changing the walls on the dead ends so that they no longer are dead ends. Dead end culling also provides the option for exceptionally easy partial braiding (with desired dead end to loop ratios). Pseudocode for dead end culling can be seen below.

- 1. Iterate through all node
- 2. If current node is a dead end (3 walls including outside borders) remove random wall excluding outside borders

(Buck Mazes for Programmers)

2.2 Pathfinding Algorithms

Pathfinding algorithms are aimed to find the shortest possible path between two set points. It has many applications such as street navigation in Google Maps, video games and maze solving. There is a multitude of pathfinding algorithms. (Algfoor, Sunar and Kolivand 1-3) 2.2.1 Dijkstra's Pathfinding Algorithm

Dijkstra's algorithm expands outwards from its starting point until it meets the ending point. There is a 100 % chance that the algorithm will find a shortest path (there can be multiple shortest paths, with the same length). The illustration below shows the algorithm working on a blank grid. The blue nodes have been visited by the algorithm, and the pink and purple nodes are the start and end points respectively. (A. Patel)

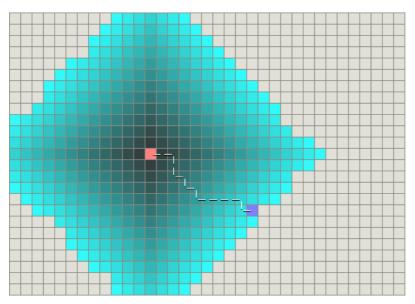


Figure 3: Image depicting Dijkstra's algorithm (A. Patel)

Pseudocode for the algorithm can be seen below.

```
// Dijkstra's Algorithm
// Set each node's position to infinity
for each node in the graph
    set the node's distance to infinity
    set the node's parent to none
// Create an unexplored set
let the unexploredSet equal a set of all the nodes
while the unexploredSet is not empty
    // Get the current node
    let the currentNode equal the node with the smallest distance
    remove the currentNode from the unexploredSet
    // Check completed
    if currentNode's position is your goal
   Congratz! You've found the end! Backtrack to get path
    // Get all the neighbors for each neighbor (still in unexploredSet) to the currentNode \,
         // Calculate the new distance
         let newDist equal currentNode's dist plus distance between
             the currentNode and the neighbor
         // Check to see if the new distance is better
         if newDist is less than currentNode's distance
             set neighbor's distance to newDist set neighbor's parent to currentNode
```

Figure 4: Pseudocode for Dijkstra's algorithm (Swift Easy Dijkstra's Pathfinding)

2.2.2 The A* Pathfinding Algorithm

A* is the most popular choice for pathfinding in video games (Mehta et al.) among others, and is a modification of Dijkstra's algorithm, and expands in the direction towards the goal. It uses a heuristic function (finding an approximate solution) to find out paths which seem to be leading to the goal and also favors paths which have the shortest path from the starting point. It always finds a shortest path. (A. Patel)

Calculating the cost of a node

The two goals of the algorithm (distance to the start and end nodes) are weighted by the f-cost, which is the overall 'cost' of a node based on its distance to the start node (g-cost) and the projected distance to the end node (h-cost). The f-, g- and h-costs are explained below.

f-cost: total cost of the node (g-cost + h-cost)

g-cost: length of the path between the node and the start

h-cost: heuristic, distance estimated to be between the node and the end. It can be acquired by using the Pythagorean theorem on the x- and y-difference between the end and current node, although other methods exist (Peters)

The pseudocode can be seen below.

```
// A* (star) Pathfinding
// Initialize both open and closed list
let the openList equal empty list of nodes
let the closedList equal empty list of nodes
// Add the start node
put the startNode on the openList (leave it's f at zero)
// Loop until you find the end
while the openList is not empty
      // Get the current node
      let the currentNode equal the node with the least f value
     remove the currentNode from the openList add the currentNode to the closedList
     // Found the goal
if currentNode is the goal
    Congratz! You've found the end! Backtrack to get path
     // Generate children
let the children of the currentNode equal the adjacent nodes
      for each child in the children
            // Child is on the closedList
           if child is in the closedList
                 continue to beginning of for loop
           // Create the f, g, and h values
child.g = currentNode.g + distance between child and current
child.h = distance from child to end
            child.f = child.g + child.h
           // Child is already in openList
if child.position is in the openList's nodes positions
   if the child.g is higher than the openList node's g
      continue to beginning of for loop
            // Add the child to the openList
            add the child to the openList
```

Figure 5: Pseudocode for the A* algorithm (Swift Easy A*)

Example illustrations of the algorithm can be seen in figures 6 and 7.

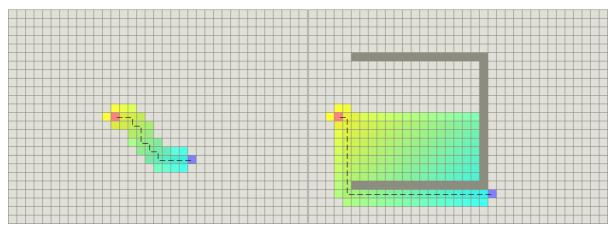


Figure 6: Unobstructed A* algorithm (A. Patel)

Figure 7: Obstructed A* algorithm (A. Patel)

2.3 Time Complexity of Algorithms

Big O notation is commonly used for the time complexity (the relation of runtime as the input gets larger). Big O notation has different cases for the best, average and worst outcomes, the worst outcome being the most used. (Cormen et al. 43-50)

Worst-Case

The worst-case complexity is done most frequently since it is easy to calculate and can show a general picture. Although it is useful, it might be too pessimistic in some cases or ignore the complete picture. (Chauan)

Best-Case

The best case shows the lower bound of the time taken for the algorithm. This isn't popular to analyze since it can't provide reliable information. An algorithm iterating over a very large data set could have small best-case time complexity, while needing years to finish operating on average. (Chauan)

Average-Case

The average case shows the time complexity of the algorithm in a more realistic and whole sense than both the worst- and best- case. It is however difficult to calculate since the 'average set of inputs' is needed to be known. The set of inputs are assessed by their probability and how much time they take, calculating the expected value. This nature of input is then used to get the time complexity. (Zeil)

2.3.1 Time complexity of Dijkstra's Algorithm

The worst-case time complexity of Dijkstra's Algorithm is $O(V^2)$, with V being the amount of vertices (nodes) in the graph. ("Shortest Path Algorithms")

To get the average case, the expected value of iterations is needed, which is wholly dependent on the input. The nature of the input is needed to be known to find the average case of Dijkstra's algorithm.

(Nilsson)

2.3.2 Time complexity of the A* algorithm

The worst-case time complexity of the A* algorithm is the same as Dijkstra's. This is due to both algorithms having to iterate through all the nodes in the worst-case, resulting in the same amount of iterations. This is also the case for the best-case, as a direct path towards the end node without any diverging paths would result in the same amount of iterations as well. However, due to using a heuristic function, the average case time complexity is aimed to be improved. (Bast)

Due to the usage of a heuristic function and the required nature of input, the average time complexity can only be determined by finding the 'quality' of the heuristic and nature of the input. (Russell and Norvig 97-104)

3. Hypothesis

Even though the best- and worst-case time complexities of Dijkstra's algorithm and A*, a heuristic function is used to make the A* algorithm be more guided towards the goal to decrease the amount of iterations, hence decreasing the overall runtime. Therefore, the A* algorithm is expected to have a shorter runtime than Dijkstra's on average.

As the size of the maze gets smaller, the cost of the heuristic function is expected to get more significant, which will result in the difference between the runtimes of the two functions getting smaller.

The average time complexities of the two algorithms can't be used to predict their runtimes because they cannot be determined without running tests on the algorithms.

4. Methodology

The recursive backtracker, dead end culling; Dijkstra's and the A* pathfinding algorithms were written for the investigation (see appendix 1). The code was written with the steps in the background information. The IDE "Visual Studio" was used for the C# implementation for the algorithms in the investigation. Although diagonal movement isn't allowed, the Pythagorean theorem is still used to calculate the heuristics function (h-cost) for the A* algorithm to maintain its integrity from real life applications.

The runtime is measured by the "System.Diagnostics.Stopwatch" class, which is the class commonly used for measuring runtime. (Allen)

The startup runtime output shows the speed at which the computer is running at the time of startup. It accesses and edits an integer variable 100000000 times.

The two pathfinding algorithms are to find the shortest path from the starting point to the ending point (declared to be at contrasting fifths of the whole grid. For example, on a 100x100 grid, the starting point is at coordinates relative to the top left corner (20,20), and the ending point at (80,80)). This allows the algorithms to venture behind the starting and ending points, rather than them being on opposite ends of the grid, not allowing any movement back.

The output of the code will be easy to transfer manually to MS Excel for analysis. Sample code output for a single trial is below.

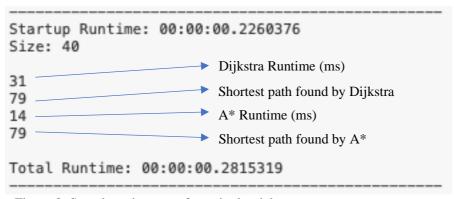


Figure 8: Sample code output for a single trial

4.1 Controlled Variables:

Maze characteristics: The characteristics of the maze are to be kept identical throughout the whole tests. Three aspects of mazes are listed below.

Elitism: Constraining the shortest path length between the start and end points could help in reducing random errors for the end result. Doing this however would be very tedious and there is no clear algorithm which improves upon the elitism.

Braiding density: The maze will not be partially braided, since randomly picking dead ends which won't be removed will increase the effect of random errors on the end result. As random errors due to the unchangeable elitism of the maze will be caused, it was opted out of having partial braiding.

Tile costs/weights: Having random tile weights (the path length between two node) would increase random errors just like the partial braiding, and therefore not implemented.

The computer that will be used (Macbook Air 2017) has the following specifications:

- 1.8GHz dual-core Intel Core i5 processor with 3MB shared L3 cache
- 8GB of 1,600MHz LPDDR3 RAM
- Intel HD Graphics 6000 (Haslam)

The exact same code except for the size of the maze (the independent variable) will be used throughout the tests.

The IDE Visual Basic and language C# will be used for the code implementation throughout the tests.

4.2 Procedure Steps

- 1. Mazes of sizes ranging from 40x40 to 320x320 with intervals of 40 (40x40, 80x80, etc.) with 10 repeats for each are generated by the recursive backtracking algorithm.
- 2. The dead-end culling algorithm is used to turn the perfect mazes generated in step 1 to braided mazes.
- 3. The mazes are solved by Dijkstra's pathfinding algorithm and the A* algorithm. The runtimes and shortest path lengths for each maze are recorded.

5. Data Presentation

After the tests, the output of the code (appendix 2) was manually translated into MS Excel, where the average and uncertainty of all the trials were calculated (raw data tables in appendix 3) then formed into the following tables.

Maze Size Against Average Shortest Path Found

The two algorithms aren't separated in this table since they had the exact same output.

Maze Size	Shortest
	Path
40x40	78 ± 20
80x80	148 ± 15
120x120	219 ± 19
160x160	287 ± 21
200x200	367 ± 33
240x240	426 ± 35
280x280	494 ± 25
320x320	559 ± 45

Table 1: Maze size against average shortest path

Maze Size Against Average Runtime of Dijkstra's and the A* Algorithm

Maze Size	Dijkstra Runtime (ms)	A* Runtime (ms)
40x40	$26,6 \pm 4,0$	$10,3 \pm 8,0$
80x80	$417,9 \pm 19,5$	$134 \pm 58,0$
120x120	2131.8 ± 60.5	$802 \pm 317,0$
160x160	$7780,9 \pm 827,5$	$2706,9 \pm 762,0$
200x200	$23049,9 \pm 2226,5$	9733,8 ± 4736,5
240x240	$51978,3 \pm 3651,0$	$24833,6 \pm 13527,0$
280x280	$109061,1 \pm 29995,0$	65271,1 ± 19982,5
320x320	$184742,9 \pm 20941,0$	$121041,2 \pm 57608,0$

Table 2: Maze size against average runtimes of Dijkstra and A*

Maze Size Against Ratio Between the Average Runtimes of Dijkstra and A*

The values were calculated simply by doing the operation Dijkstra Runtime divided by A* Runtime.

Maze Size	Ratio
40	2,58
80	3,12
120	2,69
160	2,87
200	2,37
240	2,10
280	1,67
320	1,53

Table 3: Maze size against ratio between average runtimes of Dijkstra and A*

The tables 1, 2 and 3 were used to create the following graphs (graphs 1, 2, 3, 4, 5).

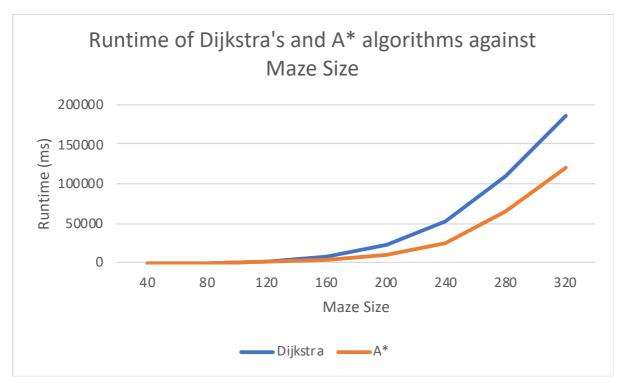
Shortest Path against Maze Size



Graph 1: Maze size against shortest path

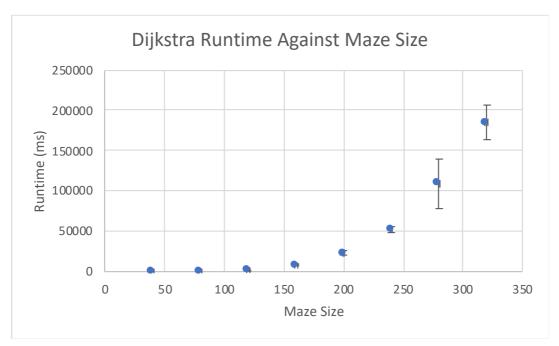
Runtime of Dijkstra's and the A* Pathfinding Algorithms against Maze Size

Uncertainties were not added to this graph as they would obstruct the view.



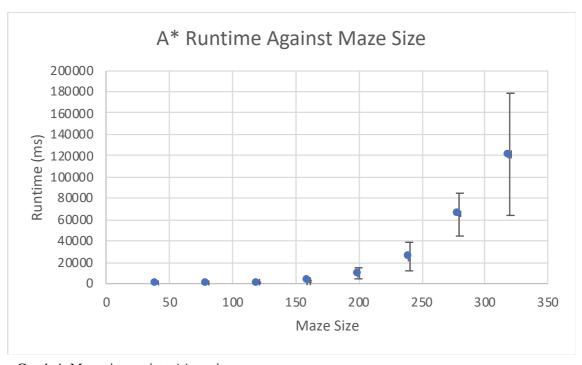
Graph 2: Maze size against Dijkstra and A* runtime

Runtime of Dijkstra's Algorithm against Maze Size



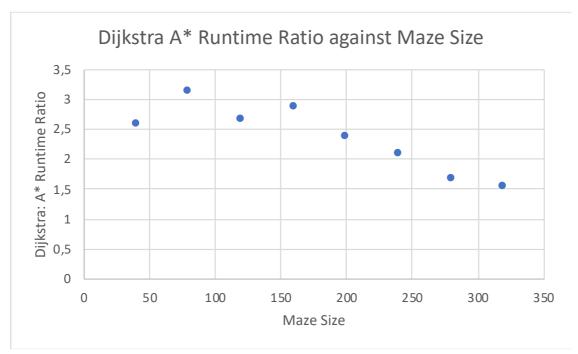
Graph 3: Maze size against Dijkstra runtime

Runtime of the A* Algorithm against Maze Size



Graph 4: Maze size against A* runtime

Dijkstra to A* Runtime Ratio against Maze Size



Graph 5: Maze size against Dijkstra to A* ratio

The graphs 1, 2, 3, 4, 5 were against the length of one axis of the maze (taken as "Maze Size") instead of its area (total number of nodes in the maze) because the shortest path length is linear to it value unlike its area. This helps in analyze the runtimes of the algorithms because they operate mainly on the shortest path and not the whole maze.

6. Data Analysis

The shortest paths found for the same mazes by the two algorithms were always the same, showing that both are capable of finding the shortest path successfully, which was predicted in the hypothesis. Seeing graph 1, the relationship between the size of one side of the maze is linear to the shortest path between the specified starting and ending points.

Seeing graph 2 and table 2, the runtime of the A* algorithm is always shorter than Dijkstra's.

It can be understood from graphs 3 and 4 that the curves roughly have the same shape (exponential increase), which relates to the fact that they derive from the same algorithm.

Seeing graph 3, Dijkstra's algorithm has unstable error bars, even while tending to have a bigger uncertainty as the maze gets larger. This is however not a direct correlation as the uncertainty for the 280x280 maze (\pm 29995) is larger than the uncertainty for the 320x320 maze (\pm 20941). These two uncertainties are also very large, showing that there are random errors in the form of the complexities of the mazes.

Seeing graph 4, the A* algorithm has its error bars widen steadily in correlation with the runtime, but the uncertainties are very large, especially at points 40 (nearly as large as the runtime value), 240 and 320 (approximately half the value). This shows that there have been a very large range of random errors throughout the tests. Despite this range of runtime, a very little portion of it falls in the range of the runtime of Dijkstra's algorithm, showing that A* has less runtime. This large range can be attributed to the predicting characteristics of the heuristic function and the varying shortest solution of the maze even when size is constant.

Finally, from the last graph it can be seen that there is a general decrease of the ratio between the runtimes of the two algorithms. This means that the difference between the two gets smaller as the size of the maze increases. This contradicts the hypothesis, as it was predicted that the difference would decrease as the size of the maze gets smaller. This change in difference might be due to the heuristic function not being able to approximate the h-cost as well in larger mazes.

From the data, it can be seen that even though their difference decreases as the maze gets larger, the A* star algorithm performs much faster that Dijkstra's algorithm. In smaller mazes such as in an 80x80 grid, it can find the same shortest path approximately 3 times faster, doing the same operation on a 320x320 grid 1,5 times faster than Dijkstra's algorithm.

7. Limitations

One of the major limitations of the methodology as seen in the data analysis is the fact that the runtime values of both algorithms are very imprecise. This is mainly caused by the fact that different mazes of the same size can have different complexities or difficulty, also having shortest paths with different lengths. This was also talked about in the control variables section, where it was stated that it was very hard to procedurally generate mazes with very similar difficulty and similar shortest path length (similar elitism). The impreciseness of the result degrades its reliability.

Another limitation which contributed to the impreciseness of the result is the fact that only 10 trials are being done for each selected maze size. This renders many different mazes untouchable and increases random errors vastly as the mazes are randomly generated. It also doesn't show the worst- and best-case scenarios.

The lack of variation of the dead ends and loops (the maze not being partially braided) might affect the results in a real-life application side. Although having random decisions between dead ends and loops would significantly increase the random errors which are already very high, they would represent a maze in a game or actual city streets much better than the current fully braided model. The same can be said about putting weights into paths between nodes.

The last graph showed that the ratio between the algorithms was decreasing but didn't show a 1:1 ratio. It cannot be known if A* will still be superior if gone into larger mazes.

8. Further Development

To increase precision, the independent variable could be changed to a specific shortest path length in a specific maze size. This could be done through setting a wanted path length and iterating until a maze which has the wanted shortest path length has been found and doing the usual tests on it. This could reduce random errors, even though the code would take significantly more time to operate.

Instead of looking at 10 samples from a specific maze size, all the mazes in that size can be evaluated. Algorithms which are uniform (can create all possible mazes) such as Wilson's or the Aldous-Broder algorithm can be used for the maze generation (Pullen). This would include the best- and worst-case scenarios in the result. This would also make finding the precise average time complexity of both algorithms possible as the nature of the input can be analyzed wholly.

Partially braiding the maze (as opposed to full braiding) or assigning weights for paths between nodes can make the maze resemble a web of streets or a videogame map more(with dead-ends, loops and harder paths to traverse), even though it would decrease the precision drastically.

9. Final Conclusion

This investigation aimed at finding the runtime difference between Dijkstra's and the A* pathfinding algorithms at solving maze problems with varying sizes. After the experiment

and result analysis, it can be concluded that the A* algorithm performs much faster than Dijkstra's algorithm.

This investigation aimed at finding the runtime difference between Dijkstra's and the A* pathfinding algorithms at solving maze problems. After doing the experiment by running the two algorithms on procedurally generated mazes of varying sizes and recording the runtimes, it can be seen that the runtime of both algorithms increased exponentially as the size (hence the shortest path) of the mazes increased. It was observed that the uncertainties of the Dijkstra algorithm increased along with the increase of the average runtime, albeit without a clear correlation. Similarly, the uncertainty of the A* algorithm increased with the average runtime, but it was consistent and much larger than Dijkstra's. The runtime of the A* algorithm was always better than Dijkstra's, with the difference between them reducing from 3 times to 1,5 times as the maze size increased.

The problem of the mazes having different lengths of shortest paths even when the size is the same (resulting in large random errors) can be solved by having the independent variable as shortest path length instead of size. Also, partially braiding the maze might lead to more realistic results or assessing all mazes for a single size can increase precision and aid in acquiring the time complexity.

Since the A* algorithm was found to be faster than Dijkstra's algorithm, it is advised to use the A* algorithm in pathfinding problems which resemble or are non-weighted mazes with multiple paths going from the start to the goal to decrease the runtime required.

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11. Appendix

11.1 Body of Code

It must be noted that the code written for the investigation isn't documented.

```
using System;
using System.Collections;
using System.Collections.Generic;
public class Node
  public int x;
  public int y;
  public Node(int xPosition, int yPosition) //argument for x,y positions in Setup()
    x = xPosition;
    y = yPosition;
  public bool isVisited = false;
  public bool[] walls = new bool[4];
  //Djkstra components
  public int dDistance = int.MaxValue;
  public Node previous = null;
  //A* components
  public int fCost;
  public int gCost;
  public int hCost;
}
namespace General_Test
  class Program
    static void Main(string[] args)
       Console.WriteLine("-----");
       var mainWatch = System.Diagnostics.Stopwatch.StartNew();
       //startup
       int iterate = 5;
       var Watch = System.Diagnostics.Stopwatch.StartNew();
       for (int i = 0; i < 100000000; i++)
         iterate++;
       Watch.Stop();
       Console.Write("Startup Runtime: ");
       Console.WriteLine(Watch.Elapsed);
```

```
//important variables
       int size = 40;
       bool braidOn = true;
       bool captionsOn = false;
       int trialAmount = 1;
       Console.Write("Size: ");
       Console.WriteLine(size);
       Console.WriteLine("");
       for (int globalCounter = 0; globalCounter < trialAmount; globalCounter++) //repeat
test
          Stack stack = new Stack();
          Node[,] allNodes = new Node[size, size];
          List<Node> FindUnvisitedNeighbors(Node inputNode)
            List<Node> neighbors = new List<Node>();
            int x = inputNode.x;
            int y = inputNode.y;
            if (x != 0)
               if (!allNodes[y, x - 1].isVisited)
                 neighbors.Add(allNodes[y, x - 1]);
            if (x != size - 1)
               if (!allNodes[y, x + 1].isVisited)
                 neighbors.Add(allNodes[y, x + 1]);
            if (y != 0)
               if (!allNodes[y - 1, x].isVisited)
                 neighbors.Add(allNodes[y - 1, x]);
            if (y != size - 1)
               if (!allNodes[y + 1, x].isVisited)
                 neighbors.Add(allNodes[y + 1, x]);
```

```
}
            return neighbors;
          }
         Node PickRandomFromNeighbors(List<Node> neighbors)
            Random rnd = new Random();
            return neighbors[rnd.Next(0, neighbors.Count)]; //can implement if statement if
0,0 doesnt work
         void RemoveWall(Node node1, Node node2)
            int xDif = node1.x - node2.x;
            int yDif = node1.y - node2.y;
            if (yDif == 1) //Top
              node1.walls[0] = true;
              node2.walls[2] = true;
            if (xDif == -1) //Right
              node1.walls[1] = true;
              node2.walls[3] = true;
            if (yDif == -1) //Bottom
              node1.walls[2] = true;
              node2.walls[0] = true;
            if (xDif == 1) //Left
              node1.walls[3] = true;
              node2.walls[1] = true;
          }
         int WallAmount(Node node)
            int wallAmount = 0;
            int x = node.x;
            int y = node.y;
            foreach (var wall in node.walls)
              //no need for checking if border node
              if (!wall)
              {
```

```
wallAmount++;
  return wallAmount;
}
void RemoveRandomWall(Node node1)
  int x = node1.x;
  int y = node1.y;
  List<Node> walledNeighbors = new List<Node>();
  if (y != 0)
    if (node1.walls[0] == false)
       walledNeighbors.Add(allNodes[y - 1, x]);
  } //top
  if (x != size - 1)
    if (node1.walls[1] == false)
       walledNeighbors.Add(allNodes[y, x + 1]);
  } //right
  if (y != size - 1)
    if (node1.walls[2] == false)
       walledNeighbors.Add(allNodes[y + 1, x]);
  } //bottom
  if (x != 0)
    if (node1.walls[3] == false)
       walledNeighbors.Add(allNodes[y, x - 1]);
  } //left
  Remove Wall (node 1, Pick Random From Neighbors (walled Neighbors)); \\
}
void Setup()
```

```
for (int i = 0; i < size; i++)
    for (int j = 0; j < size; j++)
       allNodes[i, j] = new Node(j, i);
} //Sets up the maze board
Setup();
allNodes[0, 0].isVisited = true;
stack.Push(allNodes[0, 0]);
while (stack.Count > 0)
  Node current = (Node)stack.Pop();
  List<Node> unvisitedNeighbors = FindUnvisitedNeighbors(current);
  if (unvisitedNeighbors.Count > 0)
    stack.Push(current);
    Node chosen = PickRandomFromNeighbors(unvisitedNeighbors);
    RemoveWall(current, chosen);
    chosen.isVisited = true;
    stack.Push(chosen);
  }
} //Main Maze Construction
if (braidOn)
  foreach (var node in allNodes)
    int wallAmount = WallAmount(node);
    if (wallAmount > 2)
       RemoveRandomWall(node);
} //Dead End Culling
bool flag = false;
for (int i = 0; i < size; i++)
  for (int j = 0; j < size; j++)
    bool flag2 = false;
    for (int k = 0; k < 4; k++)
```

```
{
       if (allNodes[i, j].walls[k])
         flag2 = true;
    if (!flag2)
       flag = true;
} //Maze Dimensions Output
if (!flag && captionsOn)
  Console.WriteLine("Success");
} //Checking if any node is isolated
foreach (var node in allNodes)
  node.isVisited = false;
} //reset variables
//DIJKSTRA
Node startNode = allNodes[size / 5, size / 5];
Node endNode = allNodes[4 * size / 5, 4 * size / 5]; //other end of maze
List<Node> unvisitedNodes = new List<Node>();
for (int i = 0; i < size; i++)
  for (int j = 0; j < size; j++)
    unvisitedNodes.Add(allNodes[i, j]);
} //fill unvisitedNodes
List<Node> FindUnvisitedReachableNeighbors(Node inputNode)
  List<Node> neighbors = new List<Node>();
  int x = inputNode.x;
  int y = inputNode.y;
  if (x != 0)
    if (unvisitedNodes.Contains(allNodes[y, x - 1]) && inputNode.walls[3])
       neighbors.Add(allNodes[y, x - 1]);
```

```
if (x != size - 1)
    if (unvisitedNodes.Contains(allNodes[y, x + 1]) && inputNode.walls[1])
       neighbors.Add(allNodes[y, x + 1]);
  if (y != 0)
    if (unvisitedNodes.Contains(allNodes[y - 1, x]) && inputNode.walls[0])
       neighbors.Add(allNodes[y - 1, x]);
  if (y != size - 1)
    if (unvisitedNodes.Contains(allNodes[y + 1, x]) && inputNode.walls[2])
       neighbors.Add(allNodes[y + 1, x]);
  return neighbors;
Node PickCheapestUnvisitedNode()
  int lowestCost = int.MaxValue;
  Node lowestNode = null;
  foreach (Node node in unvisitedNodes)
    if (node.dDistance < lowestCost)
       lowestCost = node.dDistance;
       lowestNode = node;
  return lowestNode;
var dijkstraWatch = System.Diagnostics.Stopwatch.StartNew();
//Dijkstra start
Node currentNode = startNode;
currentNode.dDistance = 0;
while (currentNode != endNode)
  if (currentNode == null)
    break;
```

```
List<Node> availableNeighbors =
FindUnvisitedReachableNeighbors(currentNode);
            if (availableNeighbors.Count > 0)
              foreach (Node neighbor in availableNeighbors)
                int cost = currentNode.dDistance + 1;
                if (cost < neighbor.dDistance)
                   neighbor.dDistance = cost;
                   neighbor.previous = currentNode;
              }
            }
            currentNode.isVisited = true;
            unvisitedNodes.Remove(currentNode);
            currentNode = PickCheapestUnvisitedNode();
         } //Main Body
            //Dijkstra end
         dijkstraWatch.Stop();
         if (captionsOn) { Console.Write("Dijkstra Runtime (ms): "); }
         Console.WriteLine(dijkstraWatch.ElapsedMilliseconds);
         //Shortest path output
         currentNode = endNode;
         List<Node> shortestPath = new List<Node>();
         if (endNode.previous == null)
            Console.WriteLine("No path found");
         } //path has not been found
         else
            while (currentNode != startNode)
              shortestPath.Add(currentNode);
              currentNode = currentNode.previous;
            } //Path determination
            if (captionsOn) { Console.Write("Shortest path length: "); }
            Console.WriteLine(shortestPath.Count + 1);
         } //path has been found
         currentNode = null;
         foreach (var node in allNodes)
            node.isVisited = false;
            node.dDistance = int.MaxValue;
            node.previous = null;
```

```
} //reset variables
//A STAR
List<Node> open = new List<Node>();
List<Node> closed = new List<Node>();
int visitedNodeCount = 0;
Node LowestInOpen()
  int lowestFCost = size * size;
  Node cheapestNode = open[0];
  foreach (var node in open)
    if (node.fCost < lowestFCost)
       lowestFCost = node.fCost;
       cheapestNode = node;
  if (cheapestNode != null)
    return cheapestNode;
  return null;
}
List<Node> TraversibleNotClosedNeighbors(Node inputNode)
  List<Node> neighbors = new List<Node>();
  int x = inputNode.x;
  int y = inputNode.y;
  if (x != 0)
    if (!closed.Contains(allNodes[y, x - 1]) && inputNode.walls[3])
       neighbors.Add(allNodes[y, x - 1]);
  if (x != size - 1)
    if (!closed.Contains(allNodes[y, x + 1]) && inputNode.walls[1])
       neighbors.Add(allNodes[y, x + 1]);
  if (y != 0)
```

```
if (!closed.Contains(allNodes[y - 1, x]) && inputNode.walls[0])
                                                 neighbors.Add(allNodes[y - 1, x]);
                                   if (y != size - 1)
                                         if (!closed.Contains(allNodes[y + 1, x]) && inputNode.walls[2])
                                                 neighbors.Add(allNodes[y + 1, x]);
                                   }
                                   return neighbors;
                            int CalculateFCost(Node inputNode)
                                   Node current = inputNode;
                                  while (current != startNode) //calculating path to start
                                         inputNode.gCost++;
                                         current = current.previous;
                                   }
                                   inputNode.hCost = (int)Math.Sqrt(Math.Pow(Math.Abs(inputNode.x - inputNode.hCost)) - (inputNode.hCost) -
endNode.x),2) + Math.Pow(Math.Abs(inputNode.y - endNode.y),2));
                                   return inputNode.gCost + inputNode.hCost;
                            }
                            var aStarWatch = System.Diagnostics.Stopwatch.StartNew();
                            startNode.fCost = CalculateFCost(startNode);
                            open.Add(startNode);
                            while (currentNode != endNode)
                                   currentNode = LowestInOpen();
                                   open.Remove(currentNode);
                                   closed.Add(currentNode);
                                   visitedNodeCount++;
                                   foreach (var neighbor in TraversibleNotClosedNeighbors(currentNode))
                                         if (!open.Contains(neighbor))
                                                 open.Add(neighbor);
                                                neighbor.previous = currentNode;
                                                neighbor.fCost = CalculateFCost(neighbor);
```

```
if (neighbor.gCost > currentNode.gCost + 1)
                neighbor.previous = currentNode;
                neighbor.fCost = CalculateFCost(neighbor);
           }
         }
       } //Main Body
       aStarWatch.Stop();
       if (captionsOn) { Console.Write("A* Runtime (ms): "); }
       Console.WriteLine(aStarWatch.ElapsedMilliseconds);
       if (captionsOn) { Console.Write("A* visited nodes: ");
         Console.WriteLine(visitedNodeCount);
       }
       currentNode = endNode;
       List<Node> shortestPath2 = new List<Node>();
       if (endNode.previous == null)
         Console.WriteLine("No path found");
       } //path has not been found
       else
         while (currentNode != startNode)
         {
           shortestPath2.Add(currentNode);
           currentNode = currentNode.previous;
         } //Path determination
         if (captionsOn) { Console.Write("Shortest path length: "); }
         Console.WriteLine(shortestPath2.Count + 1);
       } //path has been found
       Console.WriteLine("");
     }
    mainWatch.Stop();
    Console.Write("Total Runtime: ");
    Console.WriteLine(mainWatch.Elapsed);
    Console.WriteLine("-----");
    Console.WriteLine("");
  }
}
```

} else

10.2 Code Ouput

Startup Runtime: 00:00:00.2343633

Size: 40

Total Runtime: 00:00:00.6861862

Startup Runtime: 00:00:00.2263400

Startup Runtime: 00:00:00.2263

Size: 80

Total Runtime: 00:00:05.9816877

Startup Runtime: 00:00:00.2247201

Size: 120

Total Runtime: 00:00:30.0746075

Startup Runtime: 00:00:00.2303321

Size: 160

Total Runtime: 00:01:46.0339349

Startup Runtime: 00:00:00.2237627

Size: 200

Total Runtime: 00:05:29.5174085

Startup Runtime: 00:00:00.2262817

Size: 240

Total Runtime: 00:12:50.3569303

Startup Runtime: 00:00:00.2231192

Size: 280

Total Runtime: 00:29:06.1506214

Startup Runtime: 00:00:00.2282776

Size: 320

Total Runtime: 00:51:01.8120408

10.3 Raw Data

Maze Size		40		startup	2343633								
Dijkstra Runtim	Trial 1		Trial 2	Trial 3	Trial 4	Trial 5	Trial 6	Trial 7	Trial 8	Trial 9	Trial 10	Avg	uncertaint y
e		31	30	31	28	23	24	26	25	24	24	26,6	2
Dijkstra Path A*		87	83	103	63	75	75	81	69	79	63	77,8	20
Runtim e		19	10	16	5	10	13	13	4	10	3	10,3	
A* Path		87	83	103	63	75	75	81	69	79	63	77,8	2
Maze Size		80		startup	2263400								
Dijkstra	Trial 1		Trial 2	Trial 3	Trial 4	Trial 5	Trial 6	Trial 7	Trial 8	Trial 9	Trial 10	Avg	uncertain y
Runtim		420	418	400	406	436	404	426	405	425	439	417,9	19,
Dijkstra Path A*		137	137	135	165	151	135	151	139	163	165	147,8	1
Runtim e		95	134	144	93	88	120	188	111	204	163	134	5
A* Path		137	137	135	165	151	135	151	139	163	165	147,8	1
Maze Size		120		startup	2247201								
Dijkstra	Trial 1		Trial 2	Trial 3	Trial 4	Trial 5	Trial 6	Trial 7	Trial 8	Trial 9	Trial 10	Avg	uncertain y
Runtim													

Dijkstra													
Path A*		237	203	225	225	225	219	213	231	209	199	218,6	1
Runtim e		1163	625	811	1079	899	906	696	758	529	554	802	31
* Path		237	203	225	225	225	219	213	231	209	199	218,6	1
//aze ize		160		startup	2303321								
Dijkstra	Trial 1		Trial 2	Trial 3	Trial 4	Trial 5	Trial 6	Trial 7	Trial 8	Trial 9	Trial 10	Avg	uncertair y
tuntim		8032	8114	7047	7264	8702	8032	8205	7809	7545	7059	7780,9	827
Dijkstra Path N*		289	297	303	285	291	275	263	271	305	287	286,6	
untim		3157	3164	2379	2845	3061	3162	1938	1686	3210	2467	2706,9	7
* Path		289	297	303	285	291	275	263	271	305	287	286,6	
Лаze													
ize		200		startup	2237627								uncertai
Dijkstra Runtim	Trial 1		Trial 2	Trial 3	Trial 4	Trial 5	Trial 6	Trial 7	Trial 8	Trial 9	Trial 10	Avg	У
Dijkstra		21733	21323	21519	25714	22220	23568	25193	24566	23402	21261	23049,9	2226
ath ** tuntim		375	335	373	365	369	383	385	391	365	325	366,6	
:		7970	5700	11313	8653	9832	12346	12538	14268	9923	4795	9733,8	4736
A* Path		375	335	373	365	369	383	385	391	365	325	366,6	
Лаze iize		240		startup	2262817								
	Trial 1		Trial 2	Trial 3	Trial 4	Trial 5	Trial 6	Trial 7	Trial 8	Trial 9	Trial 10	Avg	uncertain y
Dijkstra Runtim		54796										-	
			54284	51360	51444	53293	48828	51450	48345	50336	55647	51978,3	36
Path		413	54284 465	51360 411	51444	53293 435	48828 421	51450 443	48345 395	50336 435	55647 441	51978,3 426,4	
Path N* Runtim													
Path A* Runtim		413	465	411	405	435	421	443	395	435	441	426,4	135
ath * cuntim * Path		413 22692	465 38921	411 16138	405 17293	435 28051	421 22450	443 25290	395 11867	435 29957	441 35677	426,4 24833,6	135
Path A* Runtim E A* Path		413 22692	465 38921	411 16138	405 17293	435 28051	421 22450	443 25290	395 11867	435 29957	441 35677	426,4 24833,6	135
Path * Runtim * * * * Path Maze Gize	Trial 1	413 22692 413	465 38921	411 16138 411	405 17293 405	435 28051	421 22450	443 25290	395 11867	435 29957	441 35677	426,4 24833,6	135
eath * duntim * A* Path Maze Gize Dijkstra Runtim	Trial 1	413 22692 413	465 38921 465	411 16138 411 startup	405 17293 405 2231192	435 28051 435	421 22450 421	443 25290 443	395 11867 395	435 29957 435	441 35677 441	426,4 24833,6 426,4	135 uncertain y
Path * Runtim * N* Path Maze Dijkstra Runtim Dijkstra Path Path	Trial 1	413 22692 413 280	465 38921 465 Trial 2	411 16138 411 startup Trial 3	405 17293 405 2231192 Trial 4	435 28051 435 Trial 5	421 22450 421 Trial 6	443 25290 443 Trial 7	395 11867 395 Trial 8	435 29957 435 Trial 9	441 35677 441 Trial 10	426,4 24833,6 426,4 Avg	uncertali y 299
rath * kuntim * Path A* Path Maze dize Dijkstra kuntim Silving and ath kuntim kuntim kuntim	Trial 1	413 22692 413 280	465 38921 465 Trial 2	411 16138 411 startup Trial 3	405 17293 405 2231192 Trial 4 114291	435 28051 435 Trial 5	421 22450 421 Trial 6	443 25290 443 Trial 7	395 11867 395 Trial 8	435 29957 435 Trial 9	441 35677 441 Trial 10 154838	426,4 24833,6 426,4 Avg 109061,1	uncertai y 299
ath * untim * Path Alaze ize Dijkstra cuntim bijkstra eath * untim	Trial 1	413 22692 413 280 102383 491	465 38921 465 Trial 2 103014 499	411 16138 411 startup Trial 3 102239 495	405 17293 405 2231192 Trial 4 114291 509	435 28051 435 Trial 5 110878 491	421 22450 421 Trial 6 105957 507	443 25290 443 Trial 7 102866 473	395 11867 395 Trial 8 99297 495	435 29957 435 Trial 9 94848 517	441 35677 441 Trial 10 154838 467	426,4 24833,6 426,4 Avg 109061,1 494,4	135 uncertai y 299 19982
hath * * Alaze ize bijkstra tuntim * * Path bijkstra tuntim * * * * * * * * * * * * *	Trial 1	413 22692 413 280 102383 491 75610 491	465 38921 465 Trial 2 103014 499 64205	411 16138 411 startup Trial 3 102239 495 72156 495	405 17293 405 2231192 Trial 4 114291 509 69881	435 28051 435 Trial 5 110878 491 54733	421 22450 421 Trial 6 105957 507 80189	443 25290 443 Trial 7 102866 473 40224	395 11867 395 Trial 8 99297 495 64165	435 29957 435 Trial 9 94848 517 60757	441 35677 441 Trial 10 154838 467 70791	426,4 24833,6 426,4 Avg 109061,1 494,4 65271,1	135 uncertai y 299 19982
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